

STATISTICAL MECHANICS AND ERROR-CORRECTING CODES

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Abstract: I will show that there is a deep relation between error-correction codes and certain mathematical models of spin glasses. In particular minimum error probability decoding is equivalent to finding the ground state of the corresponding spin system. The most probable value of a symbol is related to the magnetization at a different temperature. Convolutional codes correspond to one-dimensional spin systems and Viterbi's decoding algorithm to the transfer matrix algorithm of Statistical Mechanics. A particular spin-glass model, which is exactly soluble, corresponds to an ideal code, i.e. a code which allows error-free communication if the rate is below channel capacity.

The mathematical theory of communication[1, 2] is probabilistic in nature. Both the production of information and its transmission are considered as probabilistic events. A source is producing information messages according to a certain probability distribution. Each message consists of a sequence of N bits $\vec{\sigma} = \{\sigma_1, \dots, \sigma_N\}$, $\sigma_i = \pm 1$ and it is assumed that the probability $P_s(\vec{\sigma}) \equiv \exp -H_s(\vec{\sigma})$ of any particular sequence $\vec{\sigma}$ is known. According to Shannon the information content of the message is $-\ln P_s(\vec{\sigma})$ and the average information of the source is given by

$$-\sum_{\vec{\sigma}} P_s(\vec{\sigma}) \ln P_s(\vec{\sigma})$$

The messages are sent through a transmission channel. In general there is noise during transmission (which may have different origins) which corrupts the transmitted message. If a $\sigma = \pm 1$ is sent through the transmission channel, because of the noise, the output will be a real number u , in general different from σ . Again, the statistical properties of the transmission channel are supposed to be known. Let us call $Q(\vec{u}|\vec{\sigma})du$ the probability for the transmission channel's output to be between u and $u + du$, when the input was σ . $Q(\vec{u}|\vec{\sigma})$ is supposed to be known. Because of the noise during the transmission, there is a loss of information. The channel capacity \mathcal{C} is defined as the maximum information per unit time which can be transmitted through the channel. The maximum is taken over all possible sources.

Thanks to Shannon's "source coding theorem", it is always possible to encode the source in a way such that all sequences become equally probable ($H_s(\vec{\sigma}) = \text{const.}$, non depending on the σ 's). Source encoding reduces the redundancy in the source messages (not to be confused with "channel encoding", see later).

For reasons of simplicity, we will assume in the following that the source has been encoded and that the noise is independent for any pair of bits ("memoryless channel"), i.e.

$$Q(\vec{u}|\vec{\sigma}) = \prod_i Q(u_i|\sigma_i)$$

In the case of a memoryless channel and a gaussian noise, Shannon calculated the channels capacity

$$\mathcal{C} = \frac{1}{2} \log_2 \left(1 + \frac{v^2}{w^2} \right)$$

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where v^2/w^2 is the signal to noise. In the weak signal to noise limit $\mathcal{C} \sim v^2/(2w^2 \ln 2)$.

Under the above assumptions, communication is a statistical inference problem. Given the transmission channel's output and the statistical properties of the source and of the channel, one has to infer what message was sent. In order to reduce communication errors, one may introduce (deterministic) redundancy into the message ("channel encoding") and use this redundancy to infer the message sent through the channel ("decoding"). The algorithms which transform the source outputs to redundant messages are called error-correcting codes. More precisely, instead of sending the N original bits σ_i , one sends M bits J_k^{in} , $k = 1, \dots, M$, $M > N$, constructed in the following way

$$J_k^{in} = C_{i_{k_1} \dots i_{l_k}}^{(k)} \sigma_{i_{k_1}} \dots \sigma_{i_{l_k}} \quad (1)$$

where the "connectivity" matrix $C_{i_{k_1} \dots i_{l_k}}^{(k)}$ has elements zero or one. For any k , all the $C_{i_{k_1} \dots i_{l_k}}^{(k)}$ except from one are equal to zero, i.e. the J_k^{in} are equal to ± 1 . $C_{i_{k_1} \dots i_{l_k}}^{(k)}$ defines the code, i.e. it tells from which of the σ 's to construct the k th bit of the code. This kind of codes is called parity checking codes because J_k^{in} counts the parity of the minus among the l_k σ 's. The ratio $R = N/M$ which specifies the redundancy of the code, is called the rate of the code.

We illustrate with a simple example of an $R = 1/2$ code. From the N σ_i 's we construct the $2N$ $J_k^{1,in}$, $J_k^{2,in}$, $i, k = 1, \dots, N$.

$$C_{i_{k_1} i_{k_2} i_{k_3}}^{(1,k)} = \delta_{k, i_{k_1}+1} \delta_{k, i_{k_2}} \delta_{k, i_{k_3}-1}, \quad C_{i_{k_1} i_{k_2} i_{k_3}}^{(2,k)} = \delta_{k, i_{k_1}+1} \delta_{k, i_{k_3}-1}$$

$$J_k^{1,in} = \sigma_{k-1} \sigma_k \sigma_{k+1}, \quad J_k^{2,in} = \sigma_{k-1} \sigma_{k+1}$$

Knowing the source probability, the noise probability, the code and the channel output, one has to infer the message that was sent. The quality of inference depends on the choice of the code.

According to the famous Shannon's channel encoding theorem, there exist codes such that, in the limit of infinitely long messages, it is possible to communicate error-free, provided the rate of the code R is less than the channel capacity \mathcal{C} . This theorem says that such "ideal" codes exist, but does not say how to construct them.

We will now show that there exists a close mathematical relationship between error-correcting codes and theoretical models of disordered systems[3, 4, 5, 6]. As we previously said, the output of the channel is a sequence of M real numbers $\vec{J}^{out} = \{J_k^{out}, k = 1, \dots, M\}$, which are random variables, obeying the probability distribution $Q(J_k^{out} | J_k^{in})$. Once the channel output \vec{J}^{out} is known, it is possible to compute the probability $P(\vec{\tau} | \vec{J}^{out})$ for any particular sequence $\vec{\tau} = \{\tau_i, i = 1, \dots, N\}$ to be the *source* output (i.e. the information message).

More precisely, the equivalence between spin-glass models and error correcting codes is based on the following property[5, 6].

The probability $P(\vec{\tau} | \vec{J}^{out})$ for any sequence $\vec{\tau} = \{\tau_i, i = 1, \dots, N\}$ to be the information message, conditional on the channel output $\vec{J}^{out} = \{J_k^{out}, k = 1, \dots, M\}$ is given by

$$\ln P(\vec{\tau} | \vec{J}^{out}) = \text{const} - H_s(\vec{\tau}) + \sum_{k=1}^M C_{i_{k_1} \dots i_{l_k}}^{(k)} B_k \tau_{i_{k_1}} \dots \tau_{i_{l_k}} \equiv -H_t(\vec{\tau}) \quad (2)$$

where

$$B_k = B_k(J_k^{out}) = \frac{1}{2} \ln \frac{Q(J_k^{out} | 1)}{Q(J_k^{out} | -1)} \quad (3)$$

We recognize in this expression the Hamiltonian of a p-spin spin-glass Hamiltonian. The distribution of the couplings is determined by the probability $Q(J_k^{out} | J_k^{in})$.

The proof is the following. The probability $P(\vec{\tau}|\vec{J}^{out})$ for the source output to be $\vec{\tau}$ when the channel output is \vec{J}^{out} is, by Bayes formula,

$$P(\vec{\tau}|\vec{J}^{out}) = \frac{P_s(\vec{\tau})Q(\vec{J}^{out}|\vec{J}^{in})}{\sum_{\vec{\tau}} P_s(\vec{\tau})Q(\vec{J}^{out}|\vec{J}^{in})}$$

where $J_k^{in} = C_{i_{k1} \dots i_{lk}}^{(k)} \tau_{i_{k1}} \dots \tau_{i_{lk}}$. Because the channel is memoryless and $J_k^{in} = \pm 1$,

$$\ln P(\vec{\tau}|\vec{J}^{out}) = \text{const.} + \sum_{k=1}^M \ln Q(J_k^{out}|J_k^{in}) + \ln P_s(\vec{\tau})$$

$$\ln Q(J_k^{out}|J_k^{in}) = \frac{1}{2} \ln(Q(J_k^{out}|1) Q(J_k^{out}|-1)) + \frac{J_k^{in}}{2} \ln \frac{Q(J_k^{out}|1)}{Q(J_k^{out}|-1)}$$

where *const.* means independent of J^{in} . To complete the proof, one has to substitute the J^{in} 's according their definition as a product of the τ 's.

“Minimum error probability decoding” (or MED, see later), which is widely used in communications, consists in choosing the most probable sequence $\vec{\tau}^0$. This is equivalent to finding the ground state of the above spin-glass Hamiltonian.

In the case when $Q(J^{out}|J^{in}) = Q(-J^{out}|-J^{in})$ (the case of a “symmetric channel”), $B_k(J_k^{out}) = -B_k(-J_k^{out})$ and one recovers the invariance of the spin-glass Hamiltonian under gauge transformations $\tau_i \rightarrow \epsilon_i \tau_i$, $B_k \rightarrow B_k \epsilon_{i_{k1}} \dots \epsilon_{i_{lk}}$, $\epsilon_i = \pm 1$.

When all messages are equally probable and the transmission channel is memoryless and symmetric, the error probability is the same for all input sequences. It is enough to compute it in the case where all input bits are equal to one. In this case, the error probability per bit P_e is $P_e = \frac{1-m^{(d)}}{2}$, where $m^{(d)} = \frac{1}{N} \sum_{i=1}^N \tau_i^{(d)}$ and $\tau_i^{(d)}$ is the symbol sequence produced by the decoding procedure.

Let us give a couple of examples of symmetric channels. The first is the case of Gaussian noise (the “Gaussian channel”).

$$Q(J^{out}|J^{in}) = c \exp - \frac{(J^{out} - J^{in})^2}{2w^2}, \quad B_k = \frac{J_k^{out}}{w^2} \quad (4)$$

The other example is when the output is again ± 1 (the “binary symmetric channel”)

$$Q(J^{out}|J^{in}) = (1-p) \delta_{J^{out}, J^{in}} + p \delta_{J^{out}, -J^{in}}$$

$$B_k = \frac{\delta_{J_k^{out}, 1}}{2} \ln \frac{1-p}{p} + \frac{\delta_{J_k^{out}, -1}}{2} \ln \frac{p}{1-p} = \frac{J_k^{out}}{2} \ln \frac{1-p}{p} \quad (5)$$

(the last equality holds because in this case $J_k^{out} = \pm 1$).

Instead of considering the most probable instance, one may only be interested in the most probable value τ_i^p of the “bit” τ_i [7, 8, 9]. Because $\tau_i = \pm 1$, the probability p_i for $\tau_i^p = 1$ is simply related to m_i , the average of τ_i^p , $p_i = (1 + m_i)/2$.

$$m_i = \frac{1}{Z} \sum_{\{\tau_1 \dots \tau_N\}} \tau_i \exp -H_t(\vec{\tau}) \quad Z = \sum_{\{\tau_1 \dots \tau_N\}} \exp -H_t(\vec{\tau}) \quad \tau_i^p = \text{sign}(m_i) \quad (6)$$

In the previous equation m_i is obviously the thermal average at temperature $T = 1$. It is amusing to notice that for the gaussian channel or the binary symmetric channel, $T = 1$ corresponds to Nishimori’s temperature [10]. Another amusing observation is that the so-called convolutional codes which are extremely popular in communications, correspond to one dimensional spin-glasses.

Furthermore the decoding algorithm, which is called dynamical programming or “Viterbi decoding algorithm”, is nothing else than the transfer matrix algorithm of statistical mechanics.

So the equivalence between parity checking error correcting codes and theoretical models of spin glasses is quite general and we have established the following dictionary of correspondence.

<i>Error – correcting code</i>	\iff	<i>Spin Hamiltonian</i>
<i>Signal to noise</i>	\iff	$J_0^2/\Delta J^2$
<i>Maximum likelihood Decoding</i>	\iff	<i>Find a ground state</i>
<i>Error probability per bit</i>	\iff	<i>Ground state magnetization</i>
<i>Sequence of most probable symbols</i>	\iff	<i>magnetization at temperature $T = 1$</i>
<i>Convolutional Codes</i>	\iff	<i>One dimensional spin – glasses</i>
<i>Viterbi decoding</i>	\iff	<i>Transfer matrix algorithm</i>

This correspondence is not only an amusing mathematical curiosity, but can also be made useful by using the tools of modern statistical mechanics and the theory of disordered systems. Given a code, one can compute the error probability per bit if he is able to calculate the magnetization of the corresponding spin-glass model. There are at least two cases where this can be done.

a) Weak noise limit. Imagine first the case of no noise. The minimal requirement for a good code is that the corresponding spin system has a unique ground state, well separated from the excited states by a finite energy gap. Consider next slowly switching on the noise. The energy levels become random variables whose probability distribution can eventually be computed. Error occurs when there is level crossing and a formerly excited state acquires a lower energy than the spin configuration which was the ground state in the absence of noise. The probability of this to happen may be computed in certain cases.

b) Extensive connectivity. This case corresponds to a mean field limit. To be precise we consider the case of a gaussian symmetric channel (i.e. gaussian noise) and a code defined by the following connectivity matrix $C_{i_{k_1} \dots i_{l_k}}^{(k)}$ (see equation (1)); $l_k = p$ for all k and $C_{i_{k_1} \dots i_p}^{(k)} = 1$ for all possible p -spin multiplets. There are $M = N!/(p!(N-p)!)$ such multiplets. Therefore the rate of the code is $R = N/M = p!(N-p)!/(N-1)!$. We consider the limit $N \rightarrow \infty, p \rightarrow \infty, p^2/N \rightarrow 0$ and $p/\ln N \rightarrow \infty$. In this limit, the corresponding spin model is a slight generalization of Derrida’s random energy model (REM)[11]. This is easily seen, if one considers the case of all input bits equal to one. (This is not a loss of generality because all input sequences are obviously equivalent when the noise is symmetrically distributed around zero.) Derrida considered the case of gaussian random couplings with zero average and standard deviation $\Delta J^2 = W^2$. The only difference with the present case is that the coupling average is $J_0 = V$ where V^2 is the signal power which is non zero. (In fact it can be shown that a Gaussian noise is not required. Only the first two moments of the noise distribution are relevant, i.e. the computation is valid not only for Gaussian noise but also for more general symmetric noise distributions.)

For the spin model to have a nontrivial thermodynamic limit, we consider the case of a signal power such that $V = vp!/N^{p-1}$ and a noise power $W^2 = w^2p!/N^{p-1}$, p and $N \rightarrow \infty$, while v and w are kept fixed. The signal to noise power ratio is then $V^2/W^2 = v^2p!/w^2N^{p-1}$. Using arguments à la Derrida or “replica” calculations, (neither of these arguments is rigorous) it can be shown that, in this model, the ground state magnetization is $m = 1$ for $v^2/w^2 > 2 \ln 2$ and zero otherwise. As we saw above, $m = 1$ means zero error probability per bit for the corresponding code. The above inequality $v^2/w^2 > 2 \ln 2$ is equivalent to $R < C$. In other words, the error-correcting code, corresponding to the random energy model, is an ideal code, i.e. allows error-free communication if $R < C$. One may wonder how fast this code approaches the asymptotic regime. It turns out that it is possible to compute the asymptotic expansion of m as $p \rightarrow \infty$. This is done by using the

“replica method”. The result of this computation is that for $v^2/w^2 \sim 2 \ln 2$,

$$m = 1 - \frac{\exp(-pv^2/w^2)}{\sqrt{p}} \quad c(v^2/w^2) \quad c(2 \ln 2) = .987$$

To the best of my knowledge, the only other explicitly known ideal codes are pulse position modulation (or ppm) codes. They can briefly be described as follows. During a time interval T , one can transmit one of N possible symbols. T is divided into N subintervals of duration $\delta = T/N$. To send the i 'th symbol, one sends during the i 'th time subinterval an electric pulse of duration δ and amplitude h . δ and h have to be chosen depending on the noise power and the desired reliability. It can be shown that in the limit $h \rightarrow \infty$ and $\delta \rightarrow 0$, this code is ideal. Both the REM and ppm codes become ideal in the limit of infinite redundancy and zero signal to noise power.

Up to now we only considered parity checking codes, for which the “alphabet” has length 2, i.e. there are only two symbols, $\sigma = 1$ and $\sigma = -1$. Let me finally mention that many of the previous results can be generalized[5, 6] to the case of an alphabet of length l . One may establish a one to one correspondence between the l symbols of the alphabet and the elements of a finite group with the same number of elements. Spin multiplication is replaced by group multiplication. These codes can be seen as an interpolation between parity checking codes and pulse position modulation codes.

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